Reanalyses in Weather/Climate and Toward Integrated Reanalysis for Earth Systems
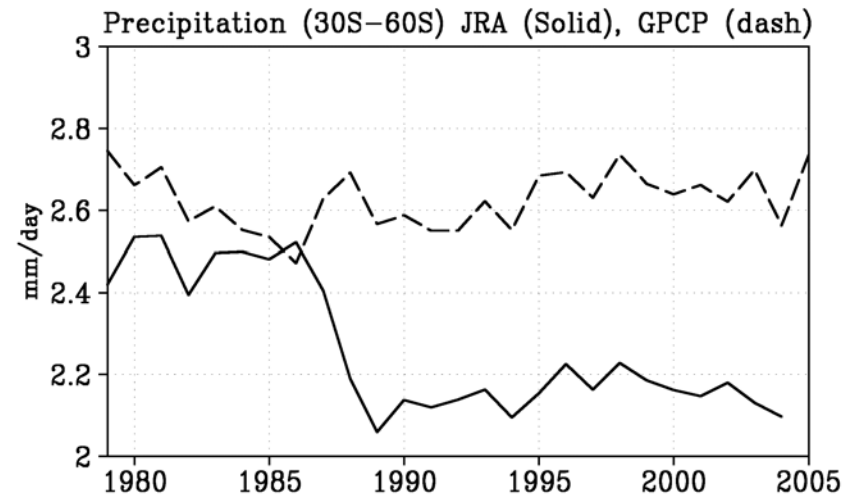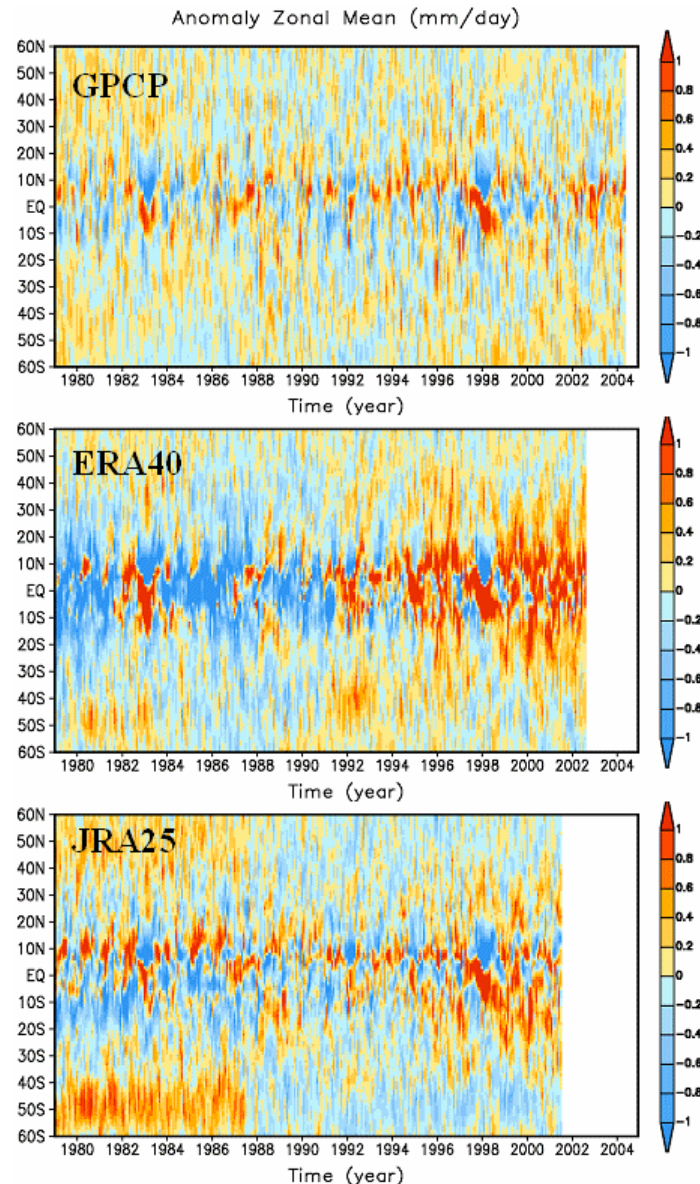
GEOS5-MERRA System Config.

- Satellite Era, 1979 – Present
- $\frac{1}{2}°\times\frac{2}{3}°$ Spatial resolution (lat-lon grid)
- 72 layers to 1 Pa
- 3 Streams (1979, 1989, 1998)
- Output Data
  - 3D Analysis (eta, prs), 6 hourly, $\frac{1}{2}°$
  - 3D Diags, prs, 3 hourly, 1.25°
  - 2D Sfc, Vert Int, Land, **1 hourly**, $\frac{1}{2}°$
  - Monthly, Monthly Diurnal, Climate Diags
  - 1.25° Coarse Resolution subset
- http://gmao.gsfc.nasa.gov/merra/

NASA and Reanalysis

- How to best integrate NASA remotely sensed data into Climate studies?
  - Is the best representation of the current state (e.g., using all of the obs.) the best analysis for climate?
- Challenges for climate reanalysis
  - The discontinuities presented by a changing observing system in data assimilation
  - Climate Model Bias
  - Uncertainty in Reanalysis
- Analysis of the Earth as a System for Climate studies:
  - Currently analyses for separate components conducted independently (e.g. Land or Ocean, using "Best" Forcing)
  - For balances, exchanges, etc, need to have consistent analyses of the separate components

Discontinuities and Trends

Anomaly Zonal Mean (mm/day)

GPCP

ERA40

JRA25

Time (year)

Precipitation (30S–60S) JRA (Solid), GPCP (dash)

- ERA40 TPW is realistic, dry model bias leads to large P trend
- JRA25 has a smaller tropical trend with sever discontinuity in the extratropics

Reanalysis Trend Differences

Air temperature (K)

Zonal Mean Circulation

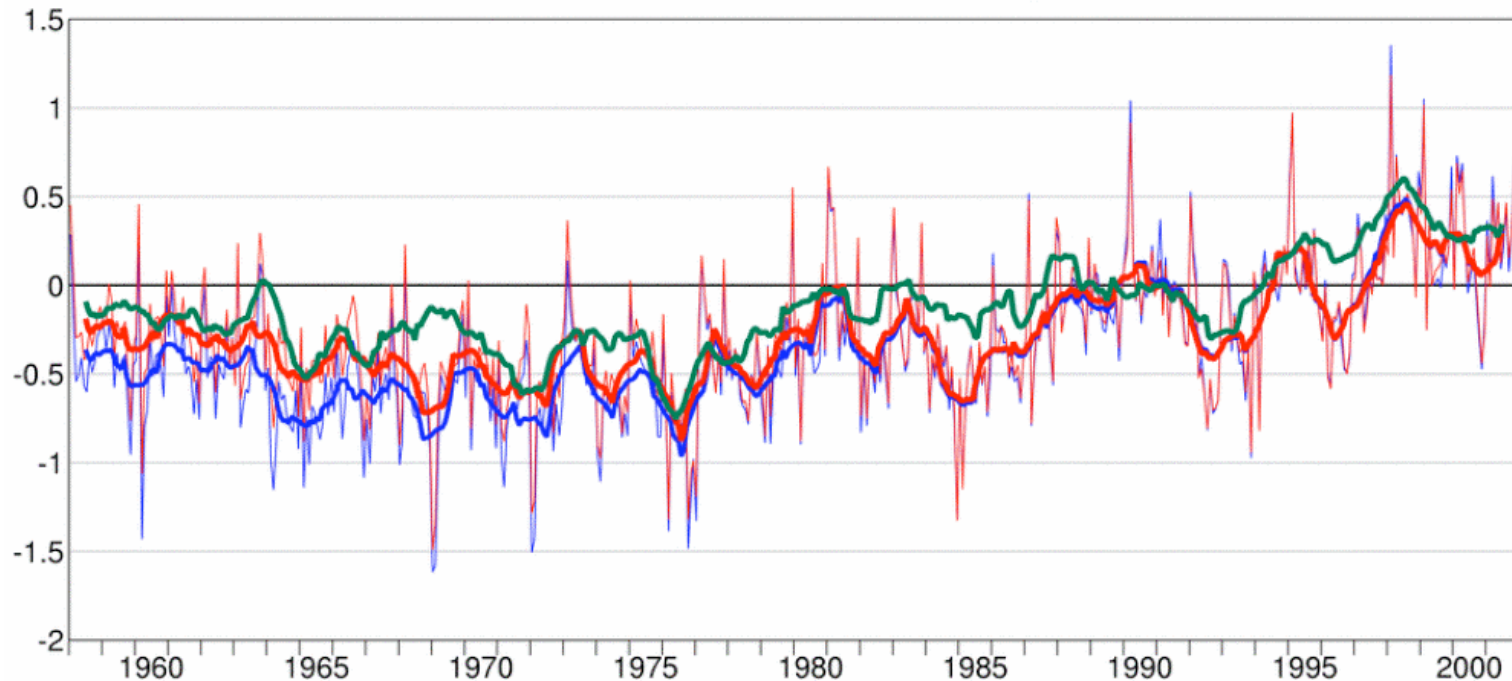10S-10N Meridional Mean Circulation
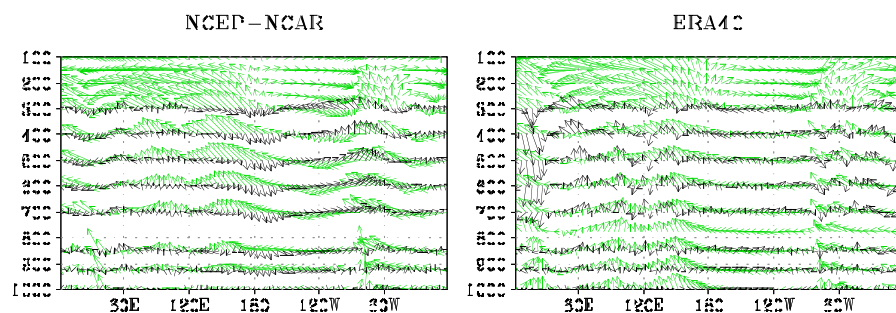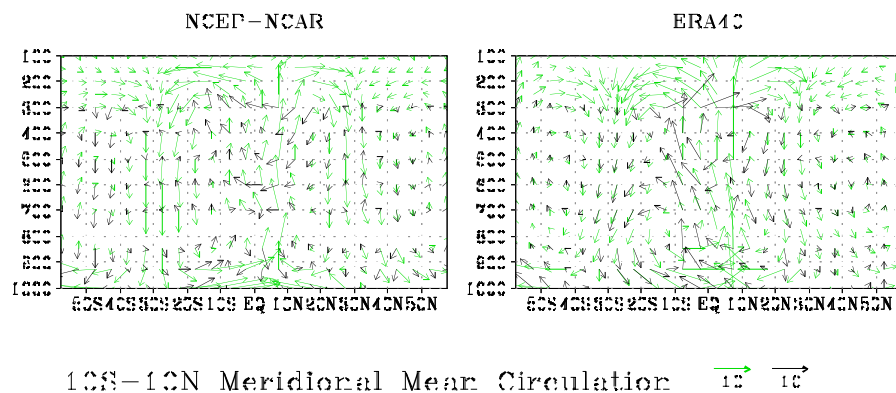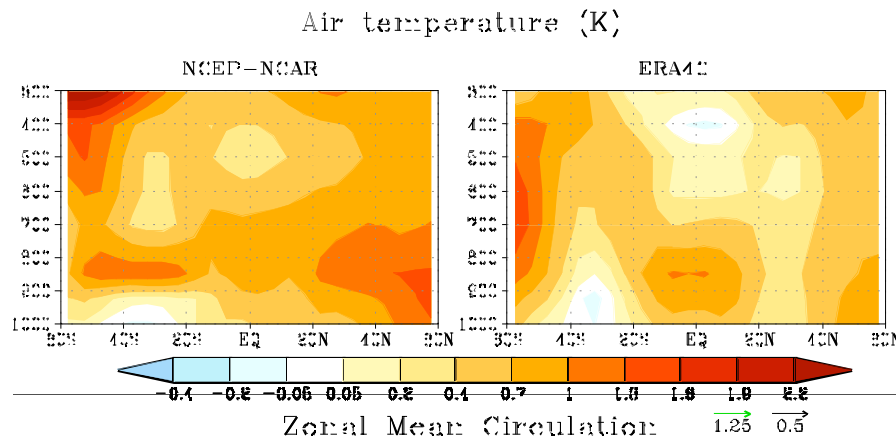
- Atmosphere Temp trends warming, with different structure

- Intensification in the Walker and Hadley circulations differ between reanalyses

Chen et al (2007, submitted)
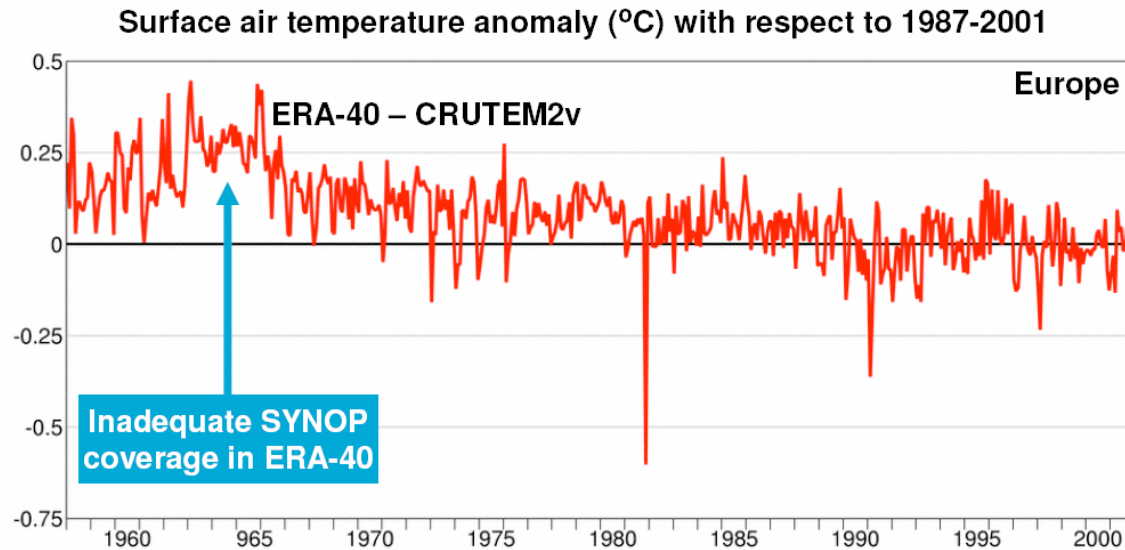
ERA40 Compared to Surface Stations (NH)

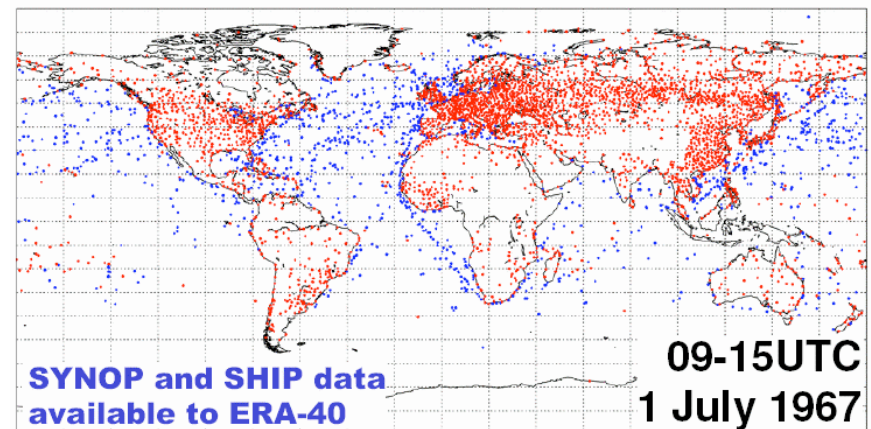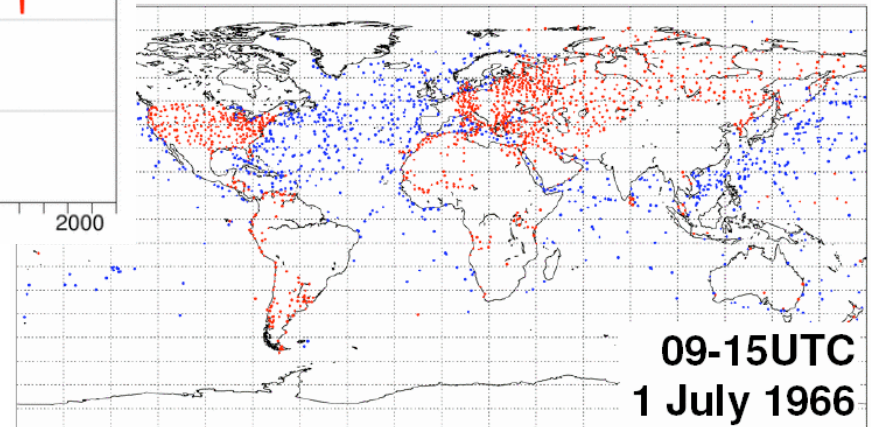**Surface air temperature anomaly (°C) with respect to 1987-2001**

Based on monthly CLIMAT data (Jones and Moberg, 2003)

Based on ERA-40 reanalysis of SYNOP data

Based on simulation using ERA-40 model and SST/sea-ice
(plotted relative to ERA-40 reanalysis mean for 1987-2001)

Simmons et al. (2004), Courtesy Dick Dee, ECMWF

Number of Observations

Surface air temperature anomaly (°C) with respect to 1987-2001

ERA-40 – CRUTEM2v

Europe

Inadequate SYNOP coverage in ERA-40

Simmons et al. 2004
Courtesy Dick Dee, ECMWF

09-15UTC
1 July 1966

SYNOP and SHIP data available to ERA-40

09-15UTC
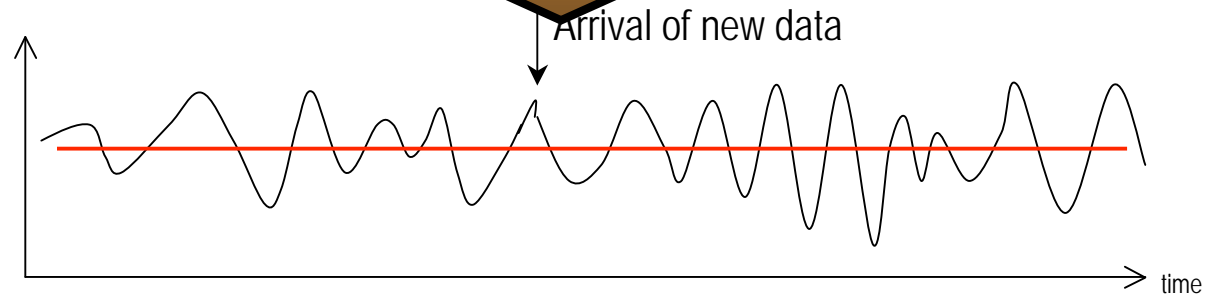1 July 1967

- New data becomes available in 1967 for ERA 40 input.

- Satellite era input data likely helps sfc temps

Climate Model Bias Correction

Analysis with a
Biased Model
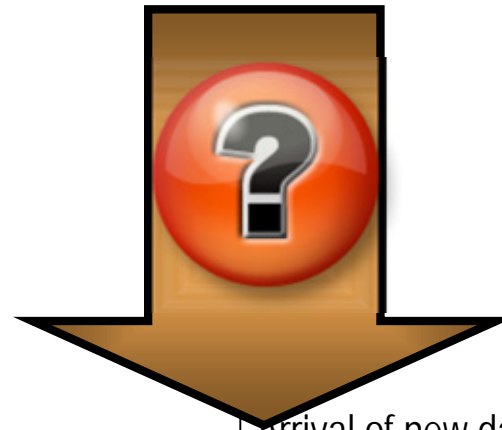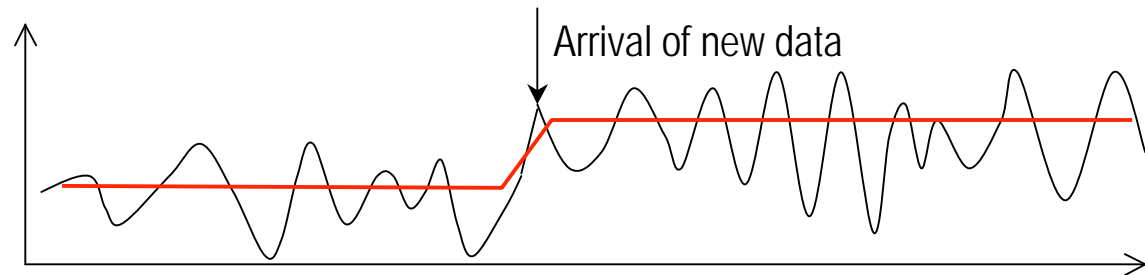
Analysis with an
Unbiased Model

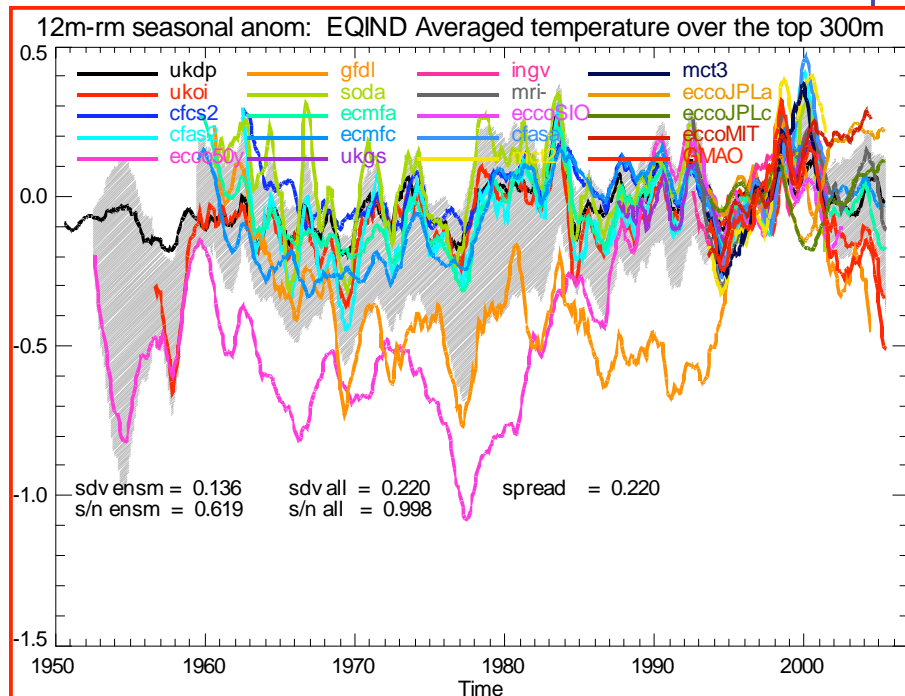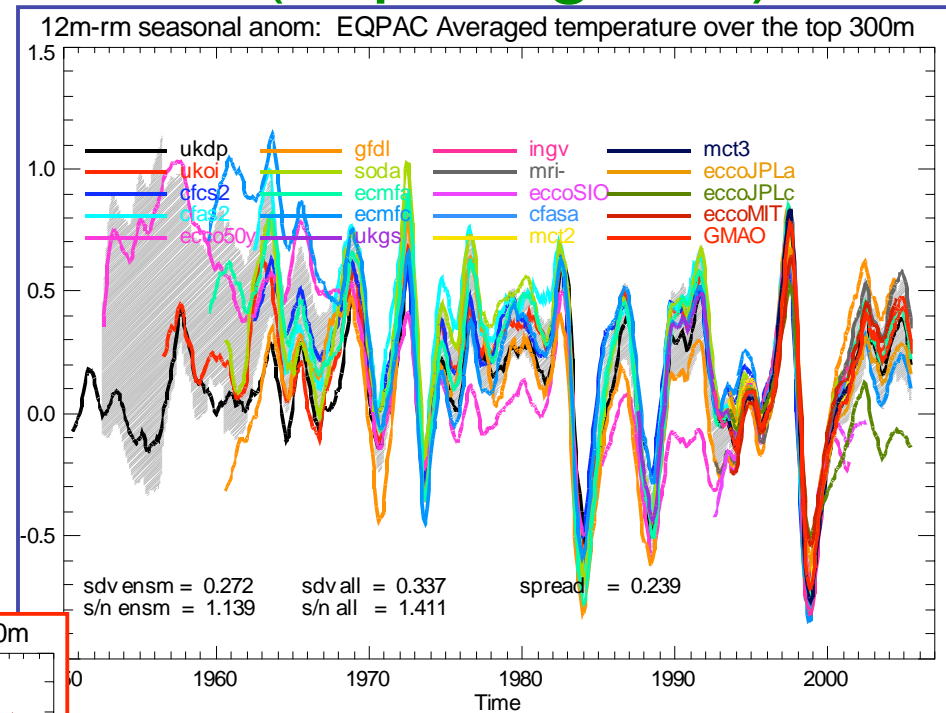Arrival of new data

Arrival of new data

time

Land Reanalysis

- ERA40 – Station meteorology analysis
- GLDAS – Offline land forcing (fluxes and met), generally not assimilating states, but using several land models (Rodell et al. 2004)
- Soil water, surface temperature, snow water all being researched (Offline – Rodell, Coupled – Reichle) for global products
- Satellite Data issues
  - Soil moisture gaps – SMMR(78-87), AMSRE (02 on), TRMM (1997 on) – Reichle et al. (2006)
  - Surface Temp requires diurnal cycle – heterogeneity and bias corrections
  - Snow water – AVHRR, SSMI data since 1980 quality of the retrievals is relatively poor

Ocean Reanalyses: T300 (Eq. Regions)

- **Eq Pac: Uncertainty decreases with time.**
- **Relatively robust interannual variability.**
- **Increased uncertainty after 2000. Why?**

12m-rm seasonal anom: EQPAC Averaged temperature over the top 300m

sdv ensm = 0.272   sdv all = 0.337   spread = 0.239
s/n ensm = 1.139   s/n all = 1.411

12m-rm seasonal anom: EQIND Averaged temperature over the top 300m

sdv ensm = 0.136   sdv all = 0.220   spread = 0.220
s/n ensm = 0.619   s/n all = 0.998

- **Eq Indian: Uncertainty remains large throughout the record.**
- **Signal to noise <1. Outliers**
- **This is also the case for the Eq Atlantic**

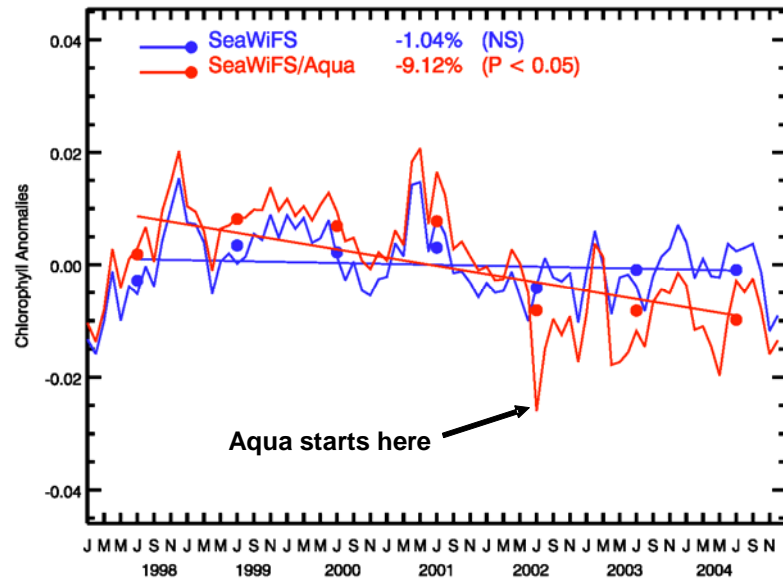# Ocean Color Analyses for Climate - the issue of consistency

Figure 2. Chlorophyll anomalies and trends for SeaWiFS (blue line), and a time series of SeaWiFS and MODIS-Aqua, with MODIS replacing SeaWiFS when it became available in late 2002 (red line). The colored dots indicate the global annual mean; the line is monthly mean. The trend for SeaWiFS is negligible. The SeaWiFS-Aqua trend is negative and statistically significant. The entire record is changed because the climatology is different when using MODIS, and it changes the anomalies.
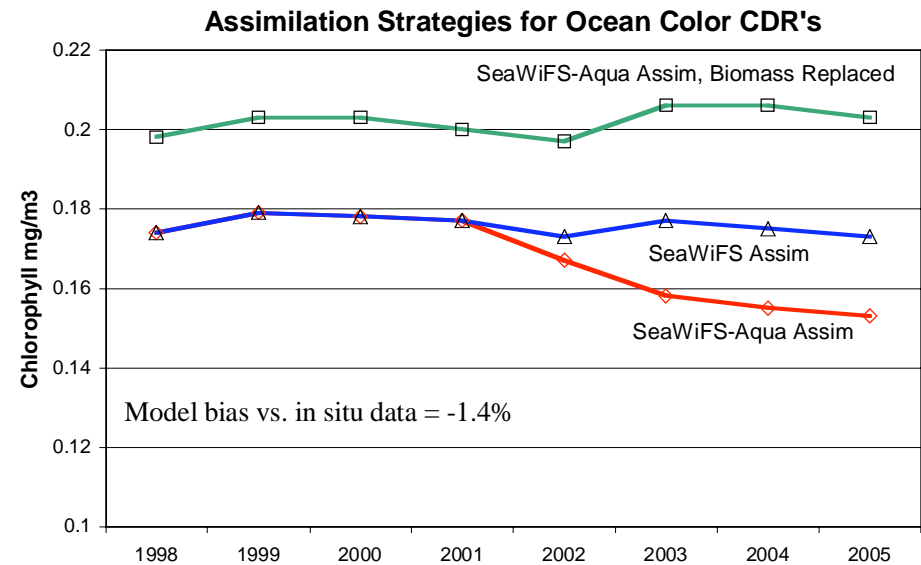

Figure 5. Trends derived from assimilation of SeaWiFS chlorophyll data (blue line), the SeaWiFS-MODIS-Aqua combination, where MODIS-Aqua replaces SeaWiFS in Sep 2002 (red line), and the bias-corrected assimilation of the SeaWiFS-MODIS-Aqua combination, where the global biomass of the satellite chlorophyll is divided out and replaced by the biomass of the model. The bias-corrected assimilation produces a chlorophyll trend that is statistically indistinguishable from the SeaWiFS assimilation. But the bias-corrected assimilation has higher global estimates of chlorophyll.

Courtesy Watson Gregg

GEOS5 Barrow, Jun/Jul 2004
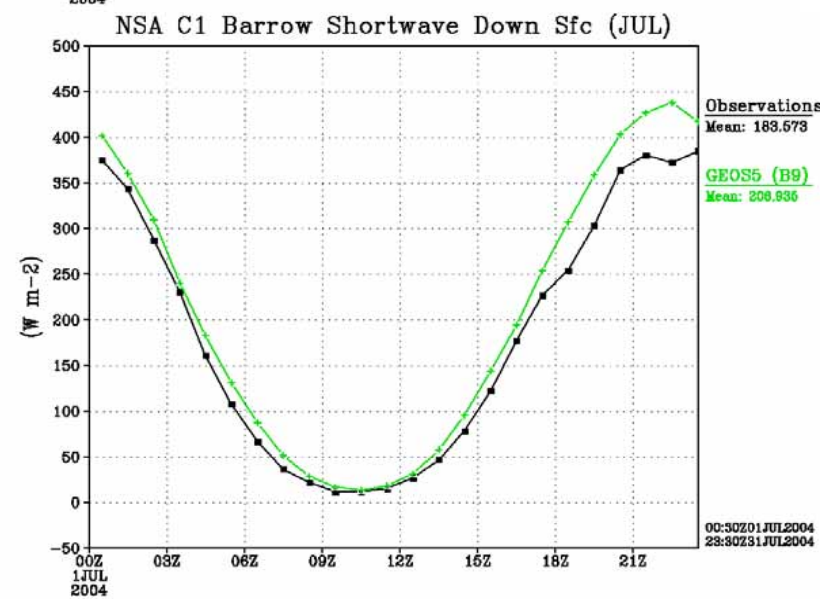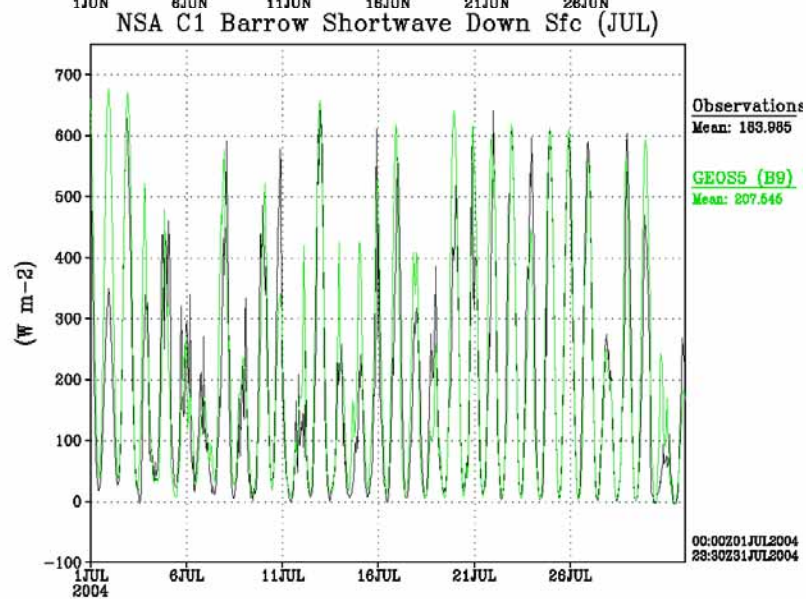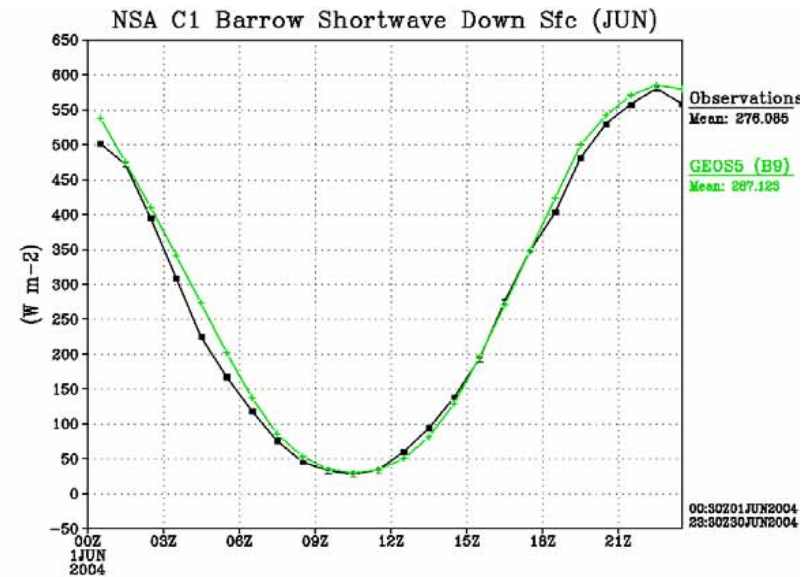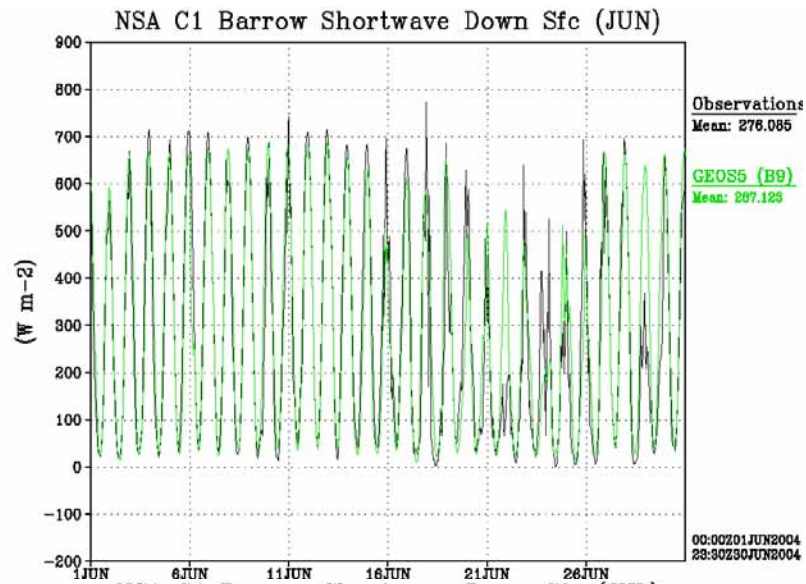
Uncertainty in Reanalyses

- Observation Inter-calibration and Homogenization
- Increments and Forecast Error
- Multi-Model evaluations (ongoing, poster)
  - 4 long reanalyses, many operational analyses
  - Observation signal should be retained in an ensemble
- Ensemble of an analysis system
  - Perturbations will affect the down stream analyses; Saturation provides some uncertainty
- Observing Systems Experiments (OSE)
  - Test the impact of entire observing systems in **reanalyses** (e.g. SSMI, AIRS) to evaluate discontinuities and trends

Extreme Diagnostics: Linking Weather and Climate

JRA25 Extreme Temp Range
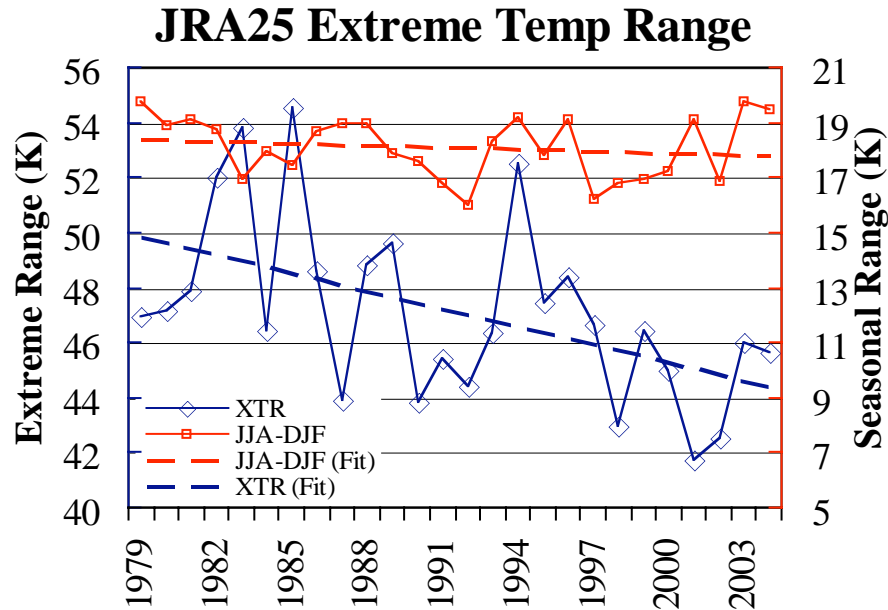
Northeastern United States area average Extreme Temperature Range (XTR, highest minus lowest daily temperature each year) and seasonal mean temperature difference (DJF minus JJA). Winter temperatures are slightly warming (not statistically significant), but there is a significant (p-value <0.01) decrease in the range of extreme temperatures.

- Tebaldi et al (2006)
- Evaluated IPCC Scenarios
  - Extreme temperature range
  - Growing season length
  - Frost days
  - Heat wave duration
  - Warm nights
  - Consecutive dry days
  - Number of heavy rain days
  - Maximum 5 day total precipitation
  - Precipitation intensity index
  - Number of rain days >95th percentile

Thanks!

(backup slides follow)
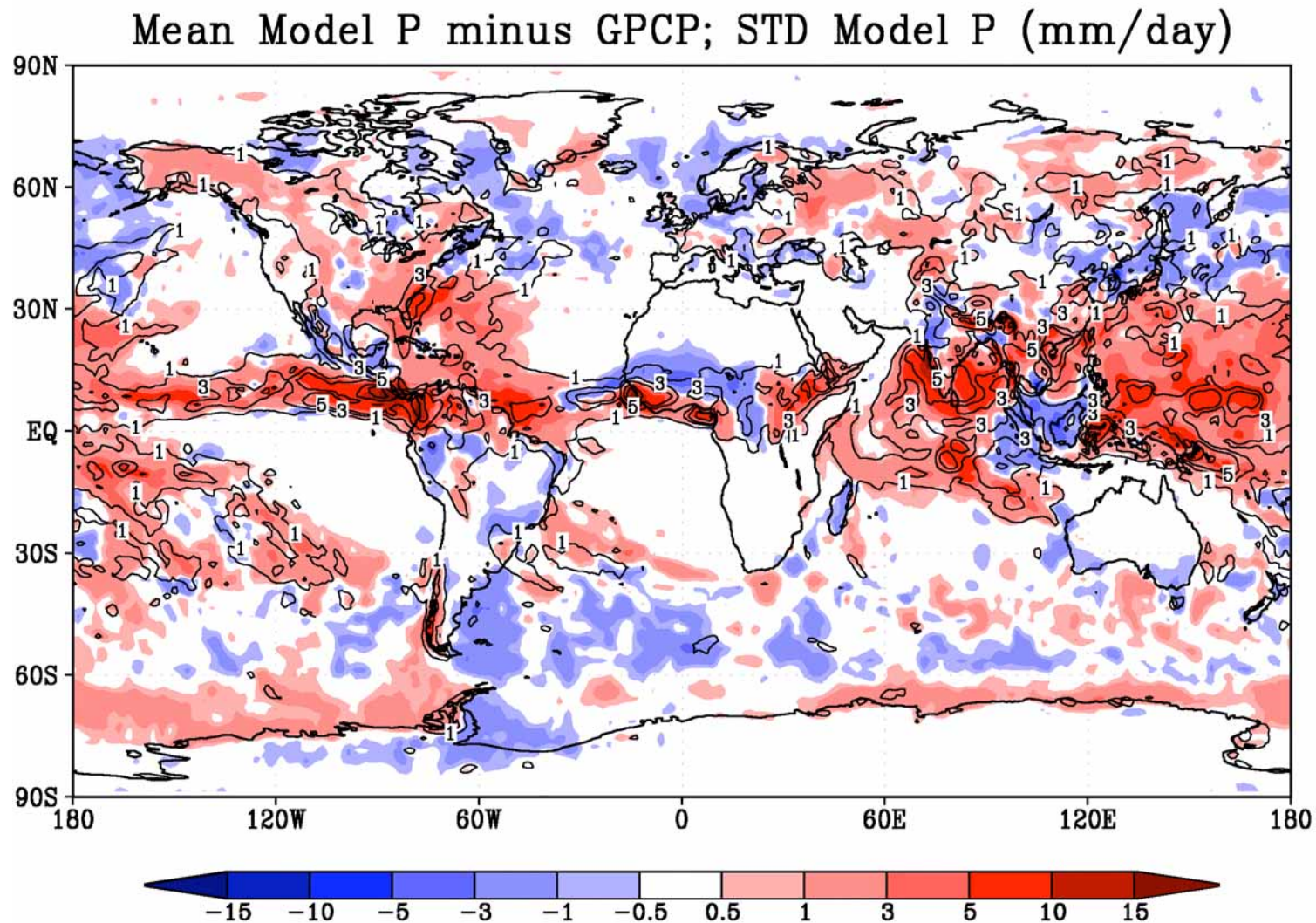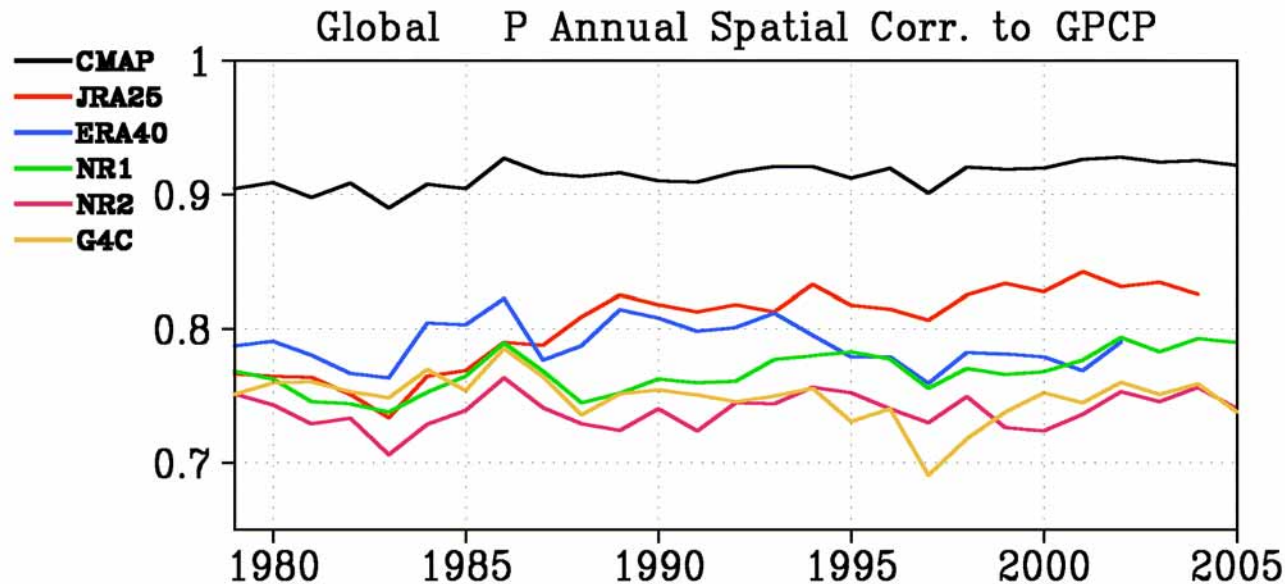
Moisture Surge during NAME 2004
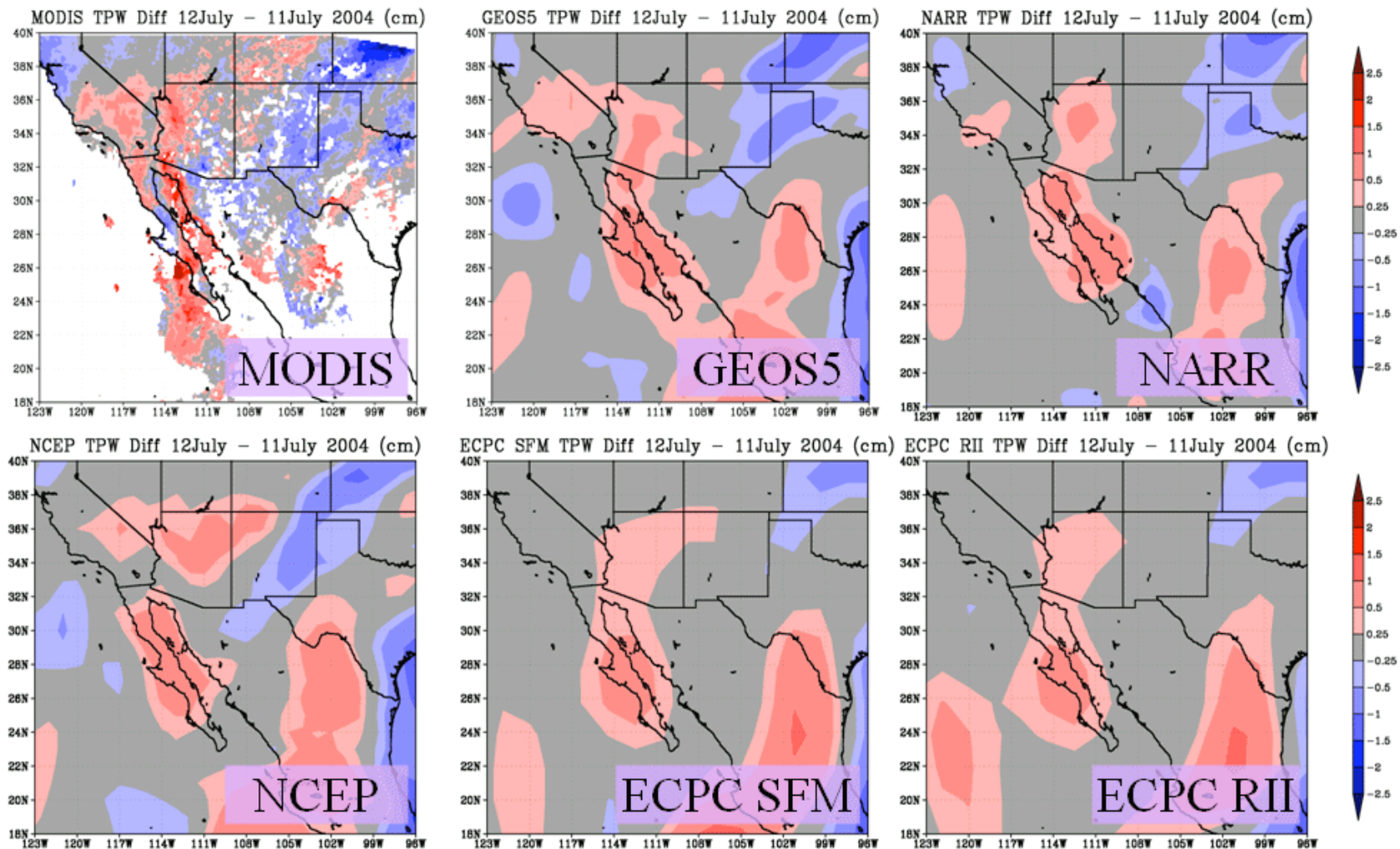
Uncertainty in Analyses

Mean Model P minus GPCP; STD Model P (mm/day)

Improving Physical Fields

- May also mean that reanalyses need to be executed and iterated more frequently than the present rate

GEOS5 Development: JJA 2004

- **Reanalyses biases are high in the tropics and scattered elsewhere**
- **CMAP Regression line is tilted off 1:1 compared with GPCP**
- **GEOS5 appears closer to CMAP**

JJA2004 Regional Precipitation

Legend: JRA25, GEOS5, NCEPR1, NCEPR2, GEOS4, GPCP, 1:1, GPCP Reg, G5 Reg
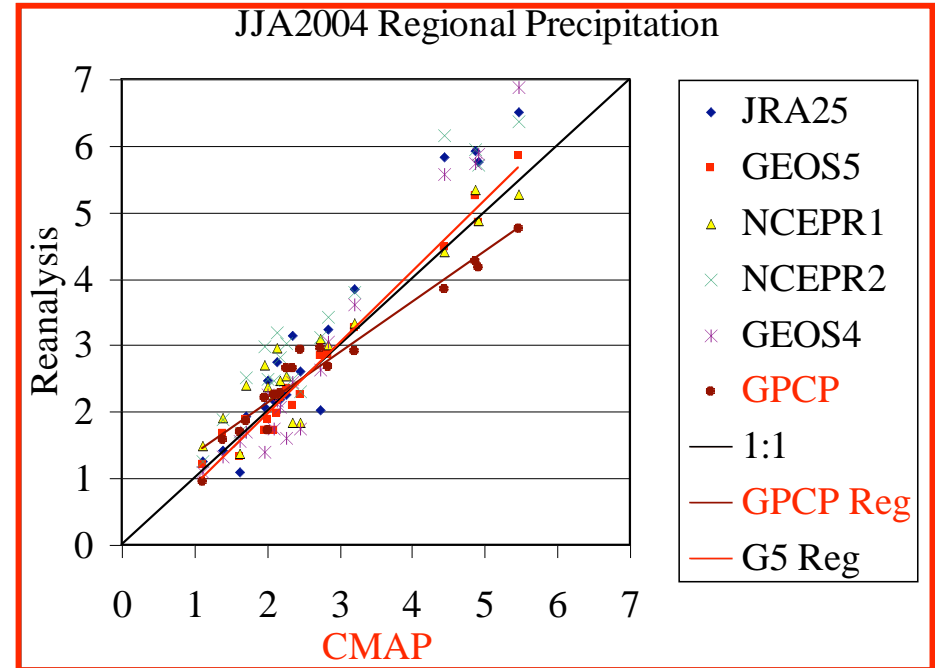
Regional average precipitation from 19 large scale oceanic and continental regions as well as global averages. The red line is the GEOS5 fit to GPCP, and the brown line is the CMAP fit.

GEOS5 Development: JJA 2004

- Spatial correlation of monthly P for global regions
- GPCP and CMAP have high land correlation
- Reanalyses need improvements especially over land and tropics
- New experiments underway

Global Precipitation Correlations to GPCP (JJA2004)

Legend: CMAP, JRA25, GEOS5, NCEPR1, NCEPR2, GEOS4, JJA-UN
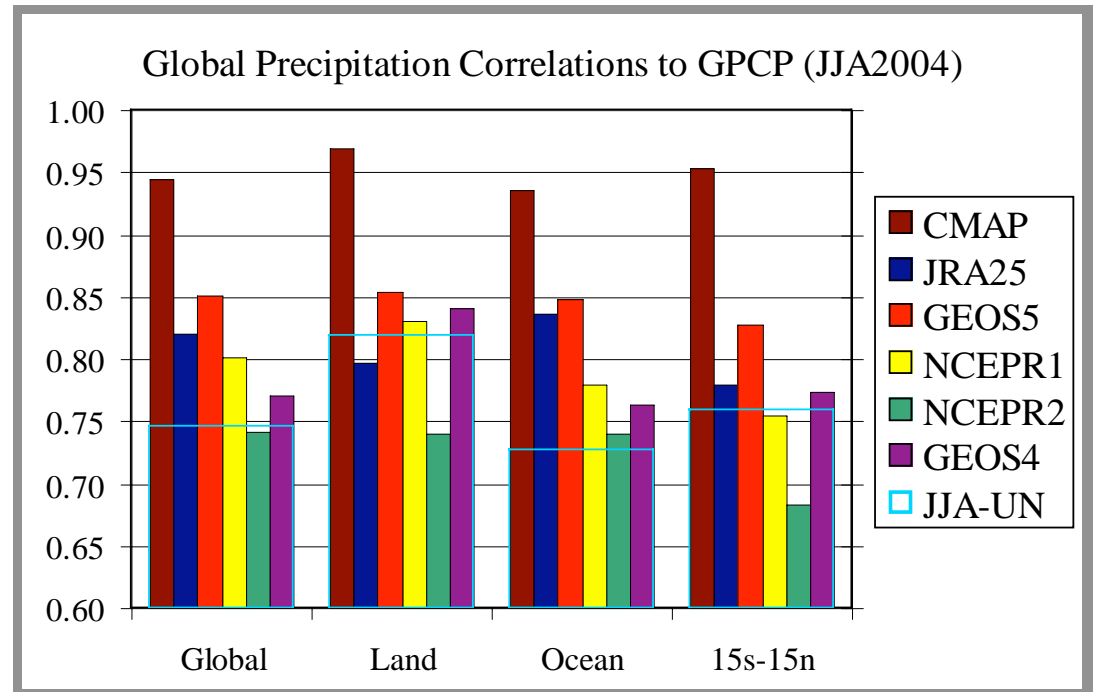
Spatial correlation of the Reanalyses systems to GPCP. The CMAP correlation (brown bar) shows the maximum expected correlation considering their uncertainties. The blue line is the average random correlation between unmatched months, a minimum of correlation.

Ocean re-analyses

- Several groups undertaking historical reanalyses

  - Scientific investigations of climate variability (CLIVAR)

  - Initial conditions for decadal prediction (e.g., Hadley Centre and collaborators in ENACT: ENhAnced ocean data assimilation and ClimaTe prediction)

  - Initial conditions for seasonal forecasts/hindcasts

- Challenge: Ocean is data-poor prior to 1993 (Topex-Poseidon, completion of TAO, Implementation of Argo program in ~2001)

- Challenge: Surface forcing products have large uncertainties/errors, particularly precipitation

- Challenge: Few observations below 750 m (Argo is changing that)

- Comparison of products at 2006 CLIVAR/GSOP Workshop at ECMWF

  - revealed a wide range in the analyses

  - some with poor comparisons with assimilated observations

  - a variety of techniques - OI, 3DVAR, reduced-state Kalman filter; 4DVAR, EnKF

- Issues: QC'd observation stream; observational error characteristics (esp. representativeness); forcing errors; background covariance models (esp. near-surface); multivariate corrections (salinity and currents are important, use of satellite observations); providing analysis uncertainties

XBT and TAO profile locations for June 1997
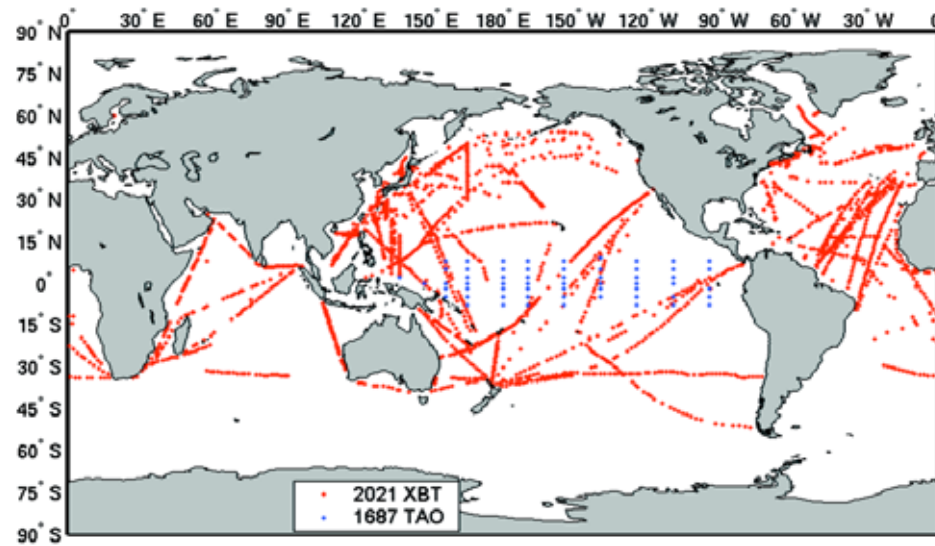
Argo profiles February 2007
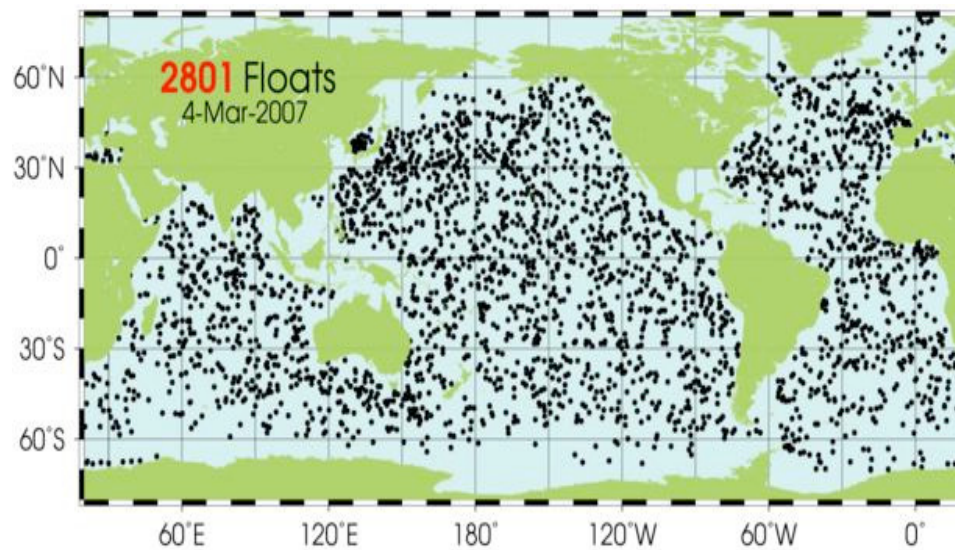
Issues in land reanalysis:

GMAO **off-line** land data assimilation development for soil moisture (mature enough for reanalysis), skin temperature (in progress), and snow (in progress).

Coupling of GMAO off-line EnKF-based land assimilation system to GEOS-5 GCM/GSI (atm. DA) in progress (not funded thus very slow).

Feedback of land assimilation on atmosphere only possible in coupled system!

Data issues:

Data gap for satellite soil moisture: retrievals ready from SMMR (1978-87) and AMSR-E (since 2002). Potential for retrievals from TRMM (since 1997) and ERS-1/2(1991-present with gaps), but not ready for reanalysis.

Skin temperature retrievals from ISCCP (multi-sensor, multi-platform, since ~1980) and MODIS (since 1999) have been used in pilot assimilation studies. A complete reanalysis requires understanding the temporal consistency, across years and across the diurnal cycle, in particular of ISCCP data. Model and observation bias issues not fully resolved.

Snow water equivalent retrievals: available since ~1980 (mostly SSM/I and AVHRR). Quality of retrievals relatively poor, more off-line testing and development needed.